Introduction – Chris Develder

- Professor at Ghent University since Oct. 2007
 - Research Interests: smart grids; dimensioning, modeling and optimizing optical (grid/cloud) networks; information retrieval/extraction
 - Visiting researcher at UC Davis, CA, USA, Jul-Oct. 2007 (optical netw.)
 - Visiting researcher at Columbia Univ., NY, USA, 2013-14 (IR/IE)
- Industry Experience: network planning/design tools
 - OPNET Technologies (now part of Riverbed), 2004-05
- PhD, Ghent University, 2003
 - "Design and analysis of optical packet switching networks"
- More info: http://users.atlantis.ugent.be/cdvelder











FACULTY OF ENGINEERING AND
ARCHITECTURE

Charging electric vehicles in the smart grid

Kevin Mets, Arun Narayanan, Matthias Strobbe, Chris Develder

Ghent University – iMinds Dept. of Information Technology – IBCN





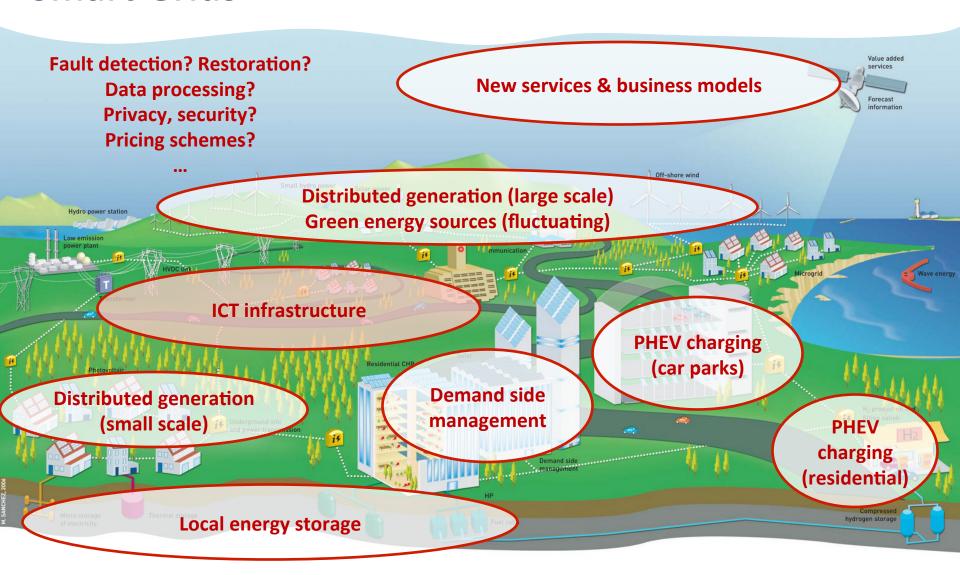
Outline

- 1. Introduction
- 2. Example 1: Peak shaving
- 3. Example 2: Wind balancing
- 4. DR algorithms for EV charging
- 5. Tools to study EV charging
- 6. Wrap-up





Smart Grids

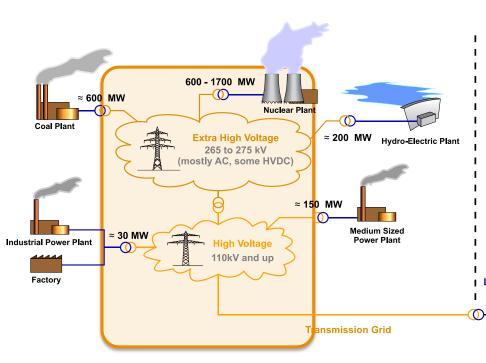




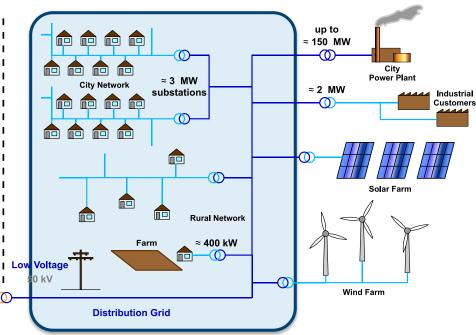


Power grid structure

Transmission network (operated by TSO)



Distribution network (operated by DSO)













Types of electrical vehicles: Key acronyms

- EV = all-electrical vehicle, akaBEV = battery electrical vehicle
- HEV = hybrid EV, includes nonelectrical motor (typically ICE)
- PEV = plug-in EV, i.e., EV or PHEV
- PHEV = plug-in hybrid EV

- ICE = internal combustion engine (i.e., burning fuel)
- EVSE = electrical vehicle supply equipment, i.e., charging station
- V2G = vehicle-to-grid = use battery to deliver power







PHEV: Toyota Prius







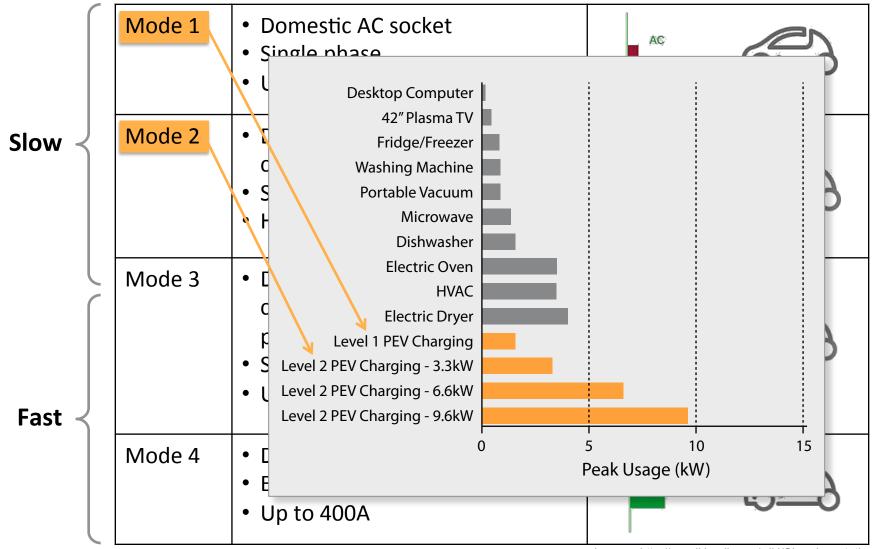
EV charging modes

	Mode 1	 Domestic AC socket Single phase Up to 16A, thus ca. 3.3kW 	AC 65-0
Slow	Mode 2	 Domestic AC socket, but special cable w/ protection Single or 3-phase Higher currents 	AC COM
Fast	Mode 3	 Dedicated AC socket, special cable w/ control (into EVSE) & protection Single or 3-phase Up to 55 kW 	Fiche spécifique
	Mode 4	DC socketExternal charger, part of EVSEUp to 400A	AC DC





EV charging modes







EV battery characteristics

	Type	Total battery capacity (kWh)	Note
Chevrolet Volt	PHEV	16.0	10.3 kWh usable (65%)
Toyota Prius Plugin	PHEV	4.4	3.4 kWh usable (77%)
Fisker Karma	PHEV	20.0	
Ford C-Max Energi	PHEV	7.6	
Volvo S60 Plugin	PHEV	12.0	
Honda Accord Plugin	PHEV	6.7	
Mitsubishi Outlander	PHEV	12.0	
Mitsubishi iMIEV	BEV	16.0	
Nissan Leaf	BEV	24.0	$20\sim22 \text{ kWh usable } (80\sim90\%)$
Smart electric drive	BEV	16.5	
Volvo C30 DrivE Electric	BEV	24.0	22.7 kWh usable (94%)
Ford Focus Electric	BEV	23.0	$19\sim20$ kWh usable (85%)
BMW i3	BEV	22.0	18.8 kWh usable (85%)
BMW ActivE	BEV	32.0	
Renault Fluence ZE	BEV	22.0	
Renault Zoe	BEV	22.0	
Volkswagen e-UP!	BEV	18.0	





EV charging modes: Time to charge

Mode	Charging time for 100km of BEV range	Power supply	Voltage	Max. current
Mode 1	6-8 hours	AC – 1-phase – 3.3 kW	230 V	16 A
	2-3 hours	AC – 3-phase – 10 kW	400 V	16 A
Mode 2	3-4 hours	AC – 1-phase – 7 kW	230 V	32 A
	1-2 hours	AC – 3-phase – 22 kW	400 V	32 A
Mode 3	20-30 minutes	AC – 3-phase – 43 kW	400 V	63 A
Mode 4	20-30 minutes	DC – 50 kW	400-500 V	100-125 A
	10 minutes	DC – 120 kW	300-500 V	300-350 A







CHAdeMO

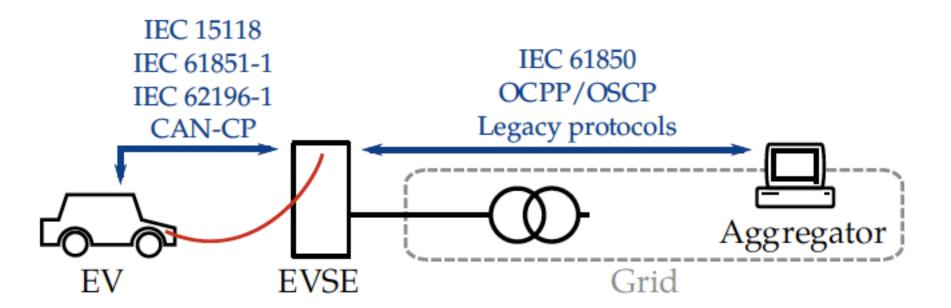


Combo



Communications?

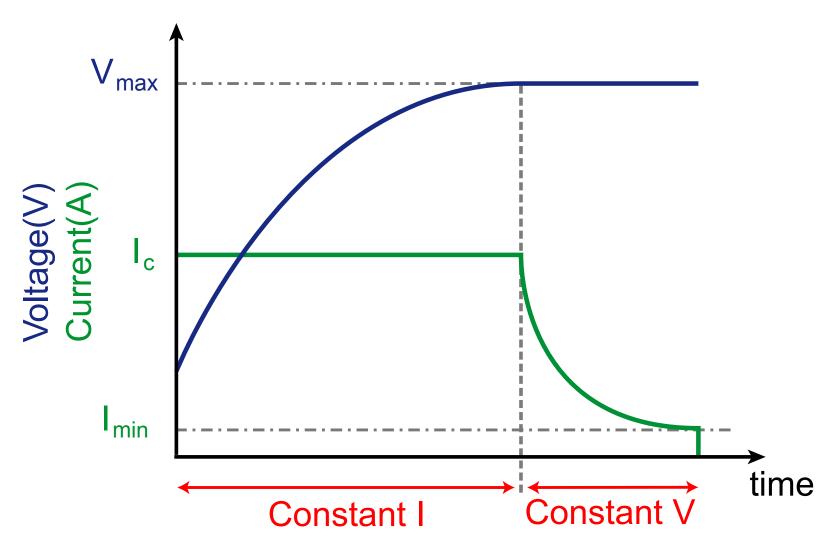
- To achieve external (= power grid) control of the charging process
- Only in ≥ Mode 2







EV charging process









Alternative charging solutions?

- Battery swapping
 - E.g., BetterPlace in Israel, Denmark
 filed for bankruptcy
 - Tesla has proprietary system



- Inductive charging
 - Volvo
 - BMW group
 - Fraunhofer
 - ...
 - Public transport initiatives







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K. Mets, R. D'hulst and C. Develder, "Comparison of intelligent charging algorithms for electric vehicles to reduce peak load and demand variability in a distribution grid", J. Commun. Netw., Vol. 14, No. 6, Dec. 2012, pp. 672-681. doi:10.1109/JCN.2012.00033







Example case study: EV charging

- Research questions:
 - 1. Impact of (uncontrolled) EV charging in a residential environment?
 - 2. Minimal impact on load peaks we could theoretically achieve?
 - 3. How can we minimize the impact of EV charging in practice?

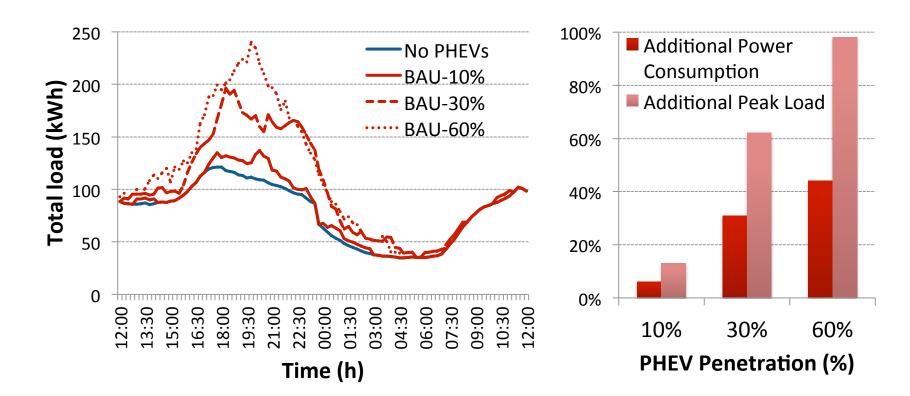






Impact of EV charging

- Sample analysis for 150 homes, x% of them own a PHEV
- BAU = maximally charge upon arrival at home



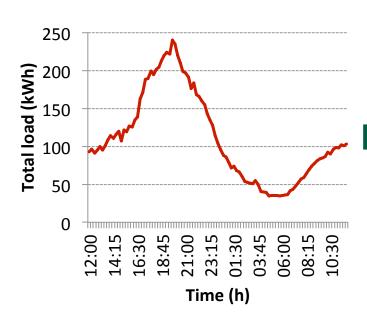


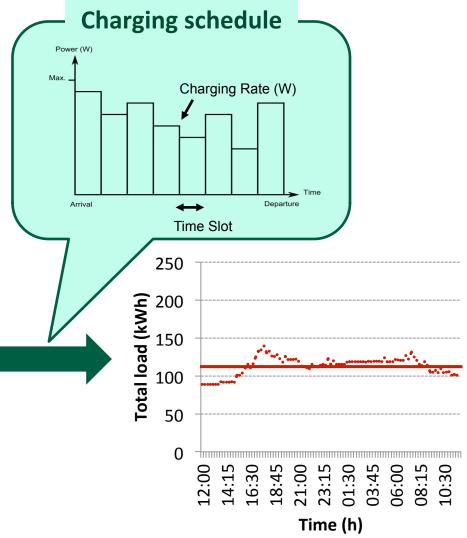


Controlling EV charging?

Objectives:

- Reduce peak load
- Flatten (total) load profile (= reduce time-variability)
- Avoid voltage violations









Smart charging algorithms

Quadratic Programming (QP)

- Offline algorithm
- Planning window
- "Benchmark"
- Three approaches:
 - Local
 - Iterative
 - Global

Multi-Agent System (MAS)

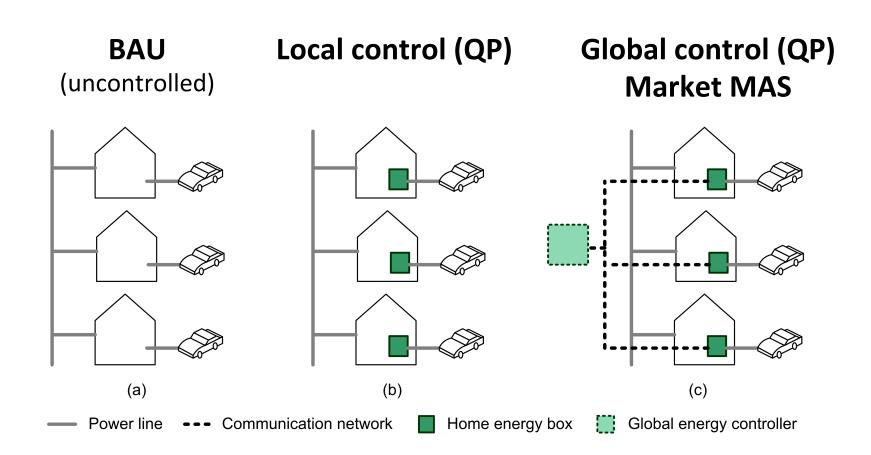
- Online algorithm
- No planning window
 → current time slot info only
 (but EV bidding changes when charging deadline approaches)
- "Realistic"
- Single approach

Reference scenario: uncontrolled charging





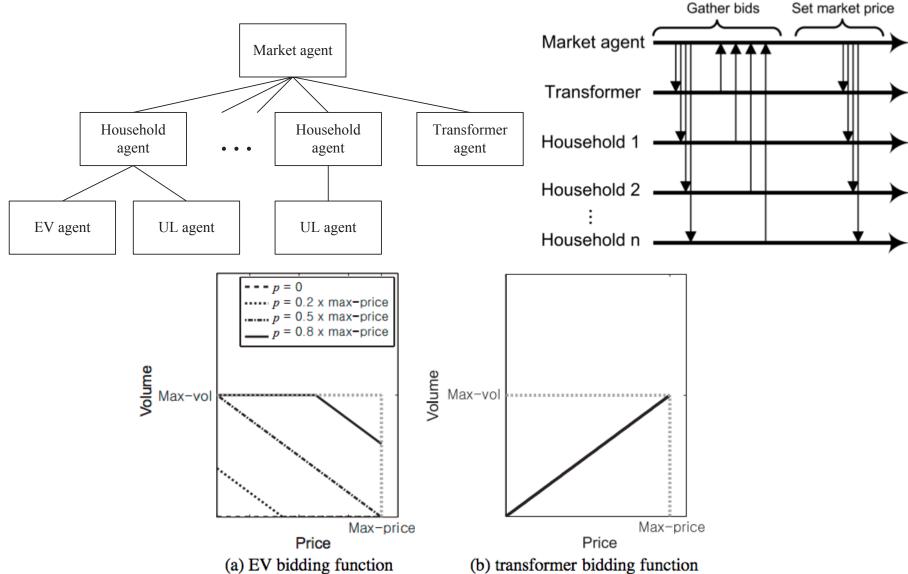
Smart charging: QP







Market-based MAS



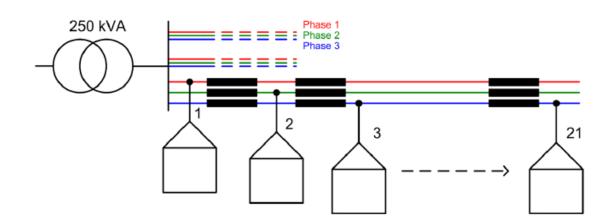




Case study

63 Households

- Randomly distributed over 3 phases
- Spread over 3 feeders



Electrical vehicles

PHEV: 15 kWh battery

Full EV: 25 kWh battery

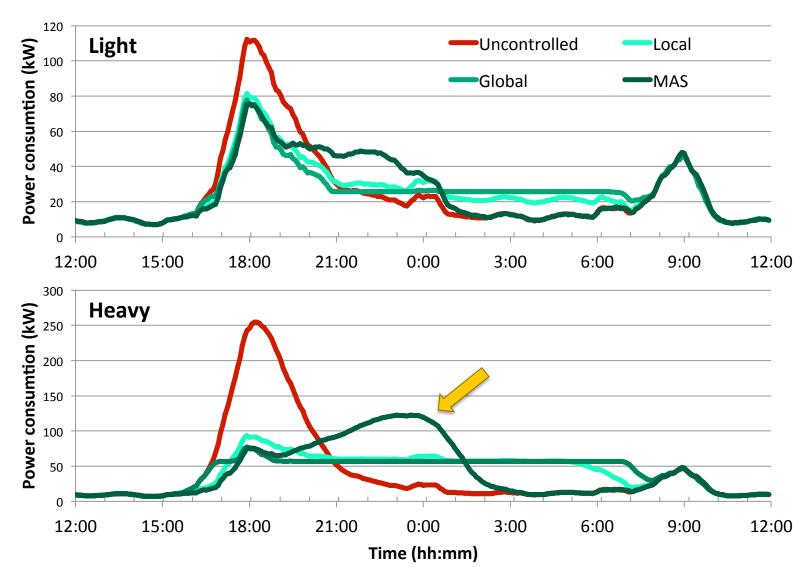
 Randomized arrivals (~5pm) and departures (~6am)

Scenario	PHEV 3.6 kW	PHEV 7.4 kW	EV 3.6 kW	EV 7.4 kW
Light	4	3	2	1
Medium	10	10	5	4
Heavy	17	16	7	7





Results (1) – Load profiles







Results (2) – Load peaks & variability

	Peak Load 📐				
Scenario	QP1 QP2		QP3	MAS	
Light	29.62%	32.16%	32.16%	32.00%	
Medium	53.84%	58.73%	58.73%	53.19%	
Heavy	63.76%	70.00%	70.00%	54.04%	

	Standard deviation \				
Scenario	QP1	QP2	QP3	MAS	
Light	35.24%	41.63%	41.94%	25.29%	
Medium	55.01%	60.50%	61.88%	34.91%	
Heavy	60.22%	63.82%	65.84%	38.80%	

$$QP1 = local$$
 $QP2 = iterative$ $QP3 = global$





Results (3) – Voltage deviations

Table 6. Average number of 5 minute time slots (out of the 288 time slots over the course of the considered one day period) during which voltage deviations

exceeding 10% are observed.

Scenario	BAU	QP1	QP2	
Light	22.17 3.90		3.31	
Medium	38.01	4.52	5.32	
Heavy	45.51	3.92	9.30	

Note: $10 \text{ slots} \sim 3.4\%$ of the time

Not solved entirely!

(No explicit part of objective function!)

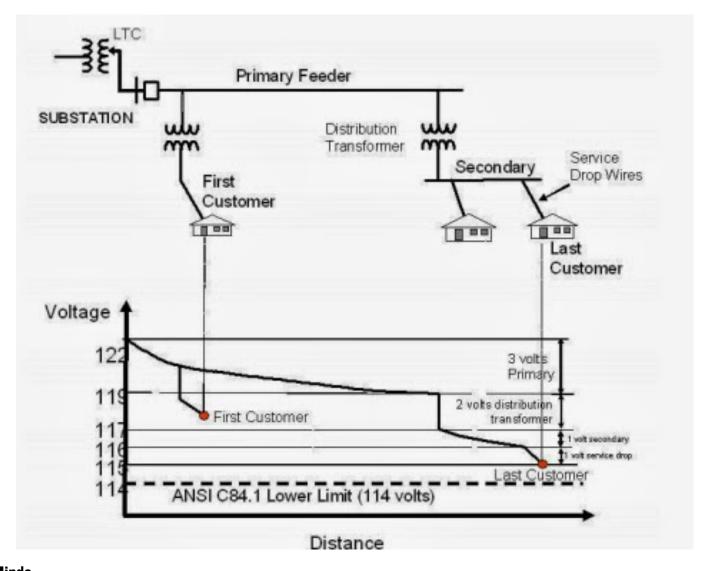
Table 7. Average and maximum magnitude of voltage deviations.

	BAU		QP1		QP2	
Scenario	AVG	MAX	AVG	MAX	AVG	MAX
Light	20%	29%	13%	19%	13%	18%
Medium	29%	60%	13%	22%	13%	20%
Heavy	37%	65%	12%	20%	14%	22%





Voltage deviation?







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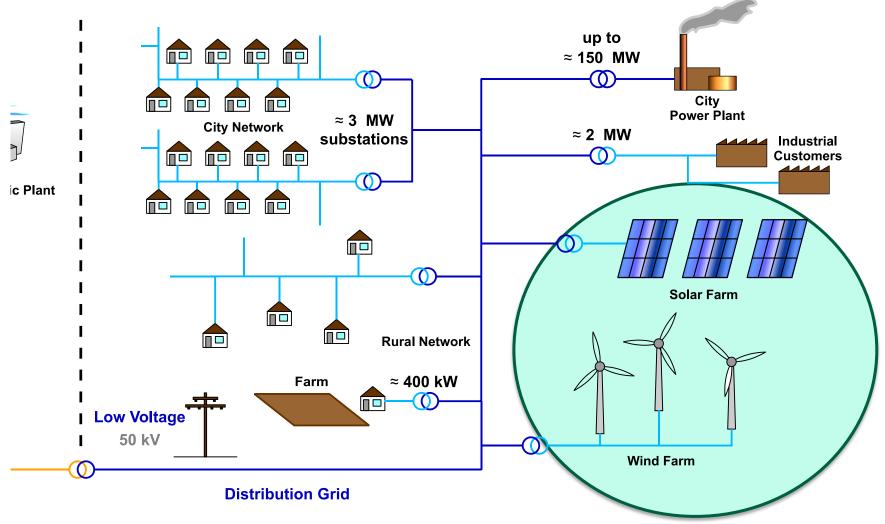
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Distributed generation (DG)







Distributed generation (DG)

Motivation for DG

- Use renewable energy sources (RES) ⇒ reduction of CO₂
- Energy efficiency, e.g., Combined Heat and Power (CHP)
- Generation close to loads
- Deregulation: open access to distribution network
- Subsidies for RES
- •

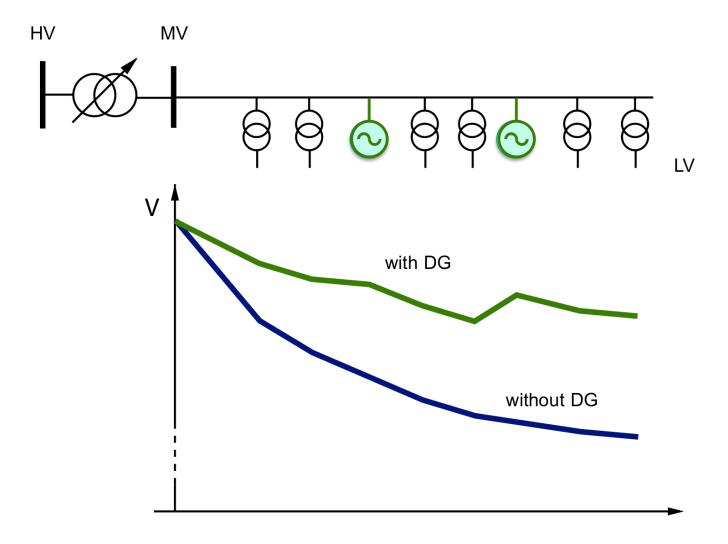
Technologies

- Wind turbines
- Photovoltaic systems
- CHP (based on fossil fuels or RES)
- Hydropower
- Biomass





Technical impact of DG?

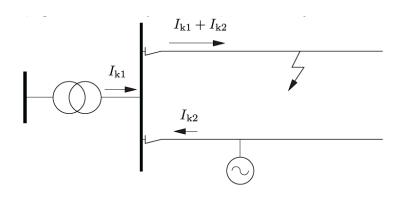






Technical impact of DG?

- Voltage variations
- Islanding?
 - Feeder disconnected from grid
 - DG may be unsafe for people & equipment
 - ...
- Power quality
 - Transient voltage variations (during connection/disconnection)
 - Cyclic variations of generator output
 - •
- Protection
 - Increase of fault currents
 - ...







Wind turbines

- Horizontal axis
 - <u>Upwind</u> vs downwind
 - Needs to be pointed into the wind
 - High rotational speed (10-22 rpm)
 - Needs a lot of space (cf. 60-90m high; blades 20-40m)



Vertical axis

- Omnidirectional
- No need to point to wind
- Lower rotational speed
- Can be closer together

E.g., http://www.inflow-fp7.eu/







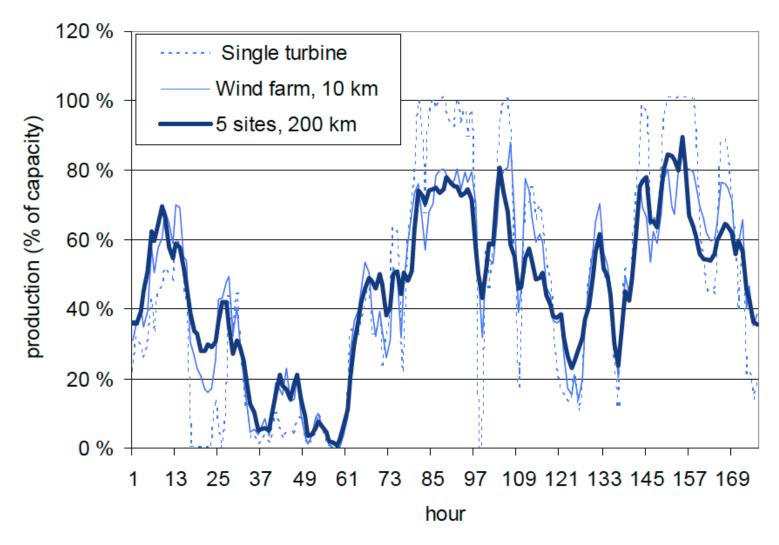
Darrieus

Savonius





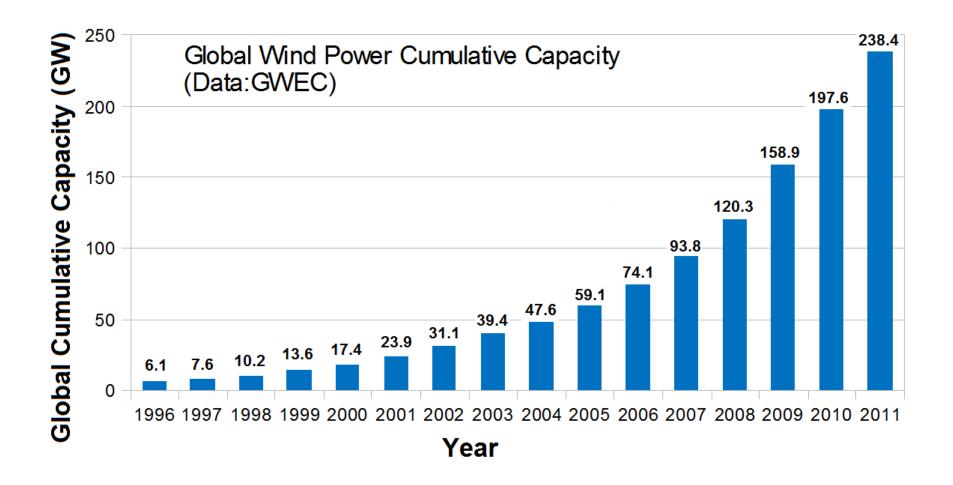
A typical wind profile







Worldwide wind power installed capacity





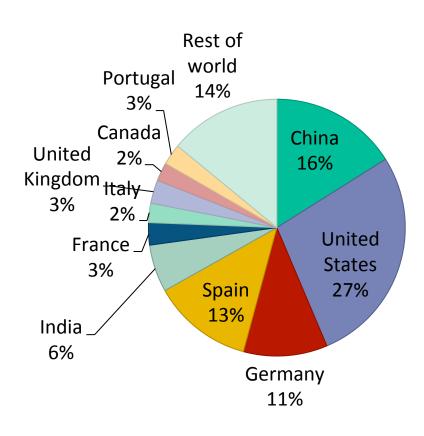


Worldwide wind power capacity & generation

Installed Capacity 2011

Canada Rest of 2%_ Portugal world 2% 13% United Kingdom China 3% 26% Italy 3% France United 3% India. **States** 7% 20% Spain 9% Germany 12%

Production 2010







Case Study

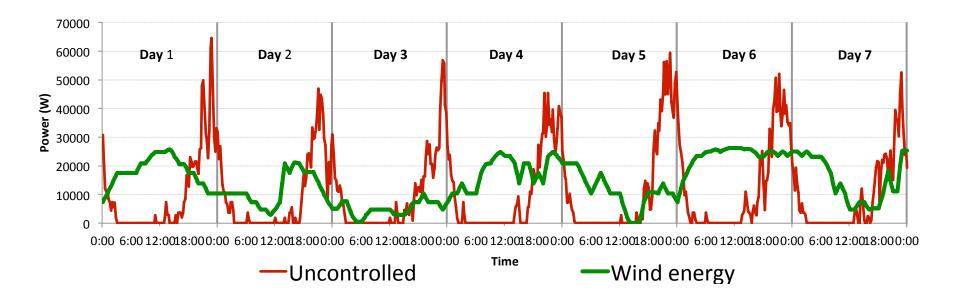
K. Mets, F. De Turck and C. Develder, "Distributed smart charging of electric vehicles for balancing wind energy", in Proc. 3rd IEEE Int. Conf. Smart Grid Communications (SmartGridComm 2012), Tainan City, Taiwan, 5-8 Nov. 2012, pp. 133-138. doi:10.1109/SmartGridComm.2012.6485972







Wind balancing



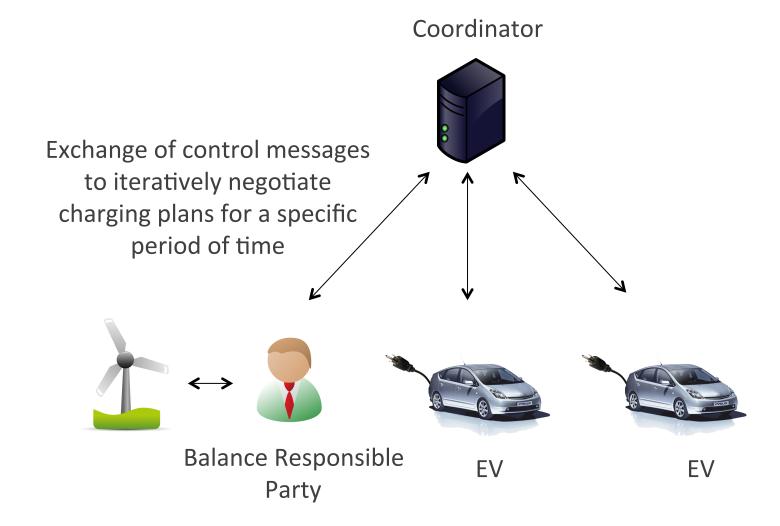
- Imbalance between supply and demand
 - Inefficient use of renewable energy sources
 - Imbalance costs
- High peak loads

Undesirable!





Architecture







Electric vehicle model

- Minimize disutility:
 - Charging schedule variables: x_t^k = charging rate for user k at time t
 - Spread demand over time, preferably at the "preferred charging rate" (p_k) , which is the maximum supported charging rate in our case
 - Model behavior/preferences of the subscriber (β_k)

$$D_t^k \left(x_t^k \right) = \beta_k^t \cdot \left(p^k - x_t^k \right)^2 \tag{1}$$

Charging schedule for a window of T time slots: minimize disutility

$$\sum_{t=1}^{I} D_t^k \left(x_t^k \right) \tag{2}$$

 $\sum_{t=1}^{T} D_t^k \left(x_t^k \right)$ Respect energy Requirement: $\sum_{t=1}^{T_k} x_t^k = E_k$ (3)

Vehicle can only be charged between arrival time S_k and departure time T_k





Balance Responsible Party Model

- Imbalance Costs
 - Minimize imbalance costs: cost penalty if supply ≠ demand
 - Supply: wind energy (w_t)
 - Demand: total of all electric vehicles (d_t)
 - Tuning parameter: α
 - Cost function: $C_t\left(d_t\right) = \alpha \cdot \left(w_t d_t\right)^2$
- For a planning window of T time slots, minimize: $\sum_{t=1}^{T} C(d_t)$





Centralized Optimization Model

- Based on social welfare maximization
 - Minimize imbalance costs
 - Minimize user disutility
- Objective: $\min_{d_t, x_t} \sum_{t=1}^T C(d_t) + \sum_{k=1}^K \sum_{t=1}^T D_t^k(x_t^k)$

Drawbacks:

- 1) Privacy: sharing of cost & disutility functions, arrival/departure info, ...
- 2) Scalability

Global constraints:

$$d_t = \sum_{k=1}^{K} x_t^k, \forall t \in \{1, 2, ..., T\}$$

- Local constraints:
 - BRP: supply < limit
 - EV: energy & time constraints





Distributed optimization model

• Move demand-supply constraint into objective, w/ Lagrange multiplier λ_t

$$\sum_{t=1}^{T} C(d_t) + \sum_{k=1}^{K} \sum_{t=1}^{T} \left(D_t^k \left(x_t^k \right) + \lambda_t \left(x_k^t - d_t \right) \right)$$

• Notice: Objective function is separable into K+1 problems that can be solved in parallel (assuming λ_t are given)

$$\begin{array}{c} \textbf{1 BRP} \\ \textbf{problem} \end{array} \underbrace{ \left[\sum_{t=1}^{T} \left(C\left(d_{t}\right) - \lambda_{t} d_{t} \right) \right]}_{T} + \underbrace{ \left[\sum_{t=1}^{K} \sum_{t=1}^{T} \left(D_{t}^{k}(x_{t}^{k}) + \lambda_{t} x_{t}^{k} \right) \right]}_{T} \underbrace{ K \text{ subscriber problems}}_{T}$$

Iteratively update pricing vector...





Distributed optimization model scheme:

- 1. Coordinator distributes virtual prices
- 2. BRP solves local problem
- 3. Subscribers solve local problem in para
- Coordinator collects schedules:
 - BRP: $d^i = [d^i_1, d^i_2, ..., d^i_T]$
 - EVs: $x^{k,i} = [x_1^{k,i}, x_2^{k,i}, ..., x_T^{k,i}]$
- 5. Coordinator updates virtual prices:

$$\lambda_t^{i+1} = \lambda_t^i + \gamma \cdot \left[\sum_{k=1}^K x_t^{k,i} - d_t^i \right]$$

6. Repeat until demand = supply





Case study: Assumptions

- Wind energy supply ≈ EV energy consumption
 - Energy supply = 6.8 MWh
- 100 Electric vehicles
 - Battery capacity: 10 kWh battery
 - Maximum charge power: 3.68 kW
 - Arrivals & departures: statistical model
 - Charging at home scenario
- Time
 - Simulate 4 weeks
 - Time slots of 15 minutes
 - Planning window of 24 hours





Case study: Algorithms

Uncontrolled business as usual (BAU)

- EV starts charging upon arrival
- EV stops charging when state-of-charge is 100%
- No control or coordination

Distributed algorithm

Executed at the start of each time slot

"Ideal world" benchmark

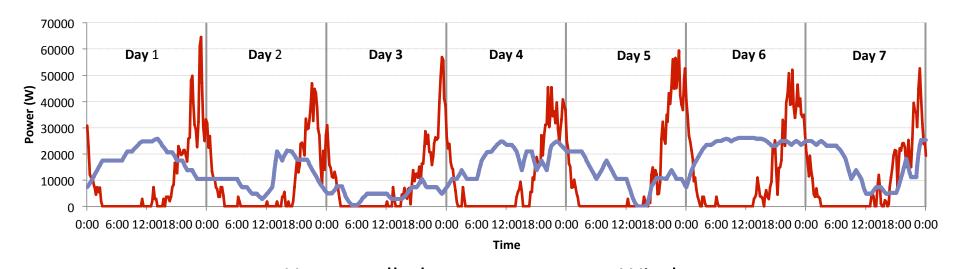
- Offline all-knowing algorithm determines schedules for ALL sessions
- No EV disutility function → maximum flexibility
- Objective:

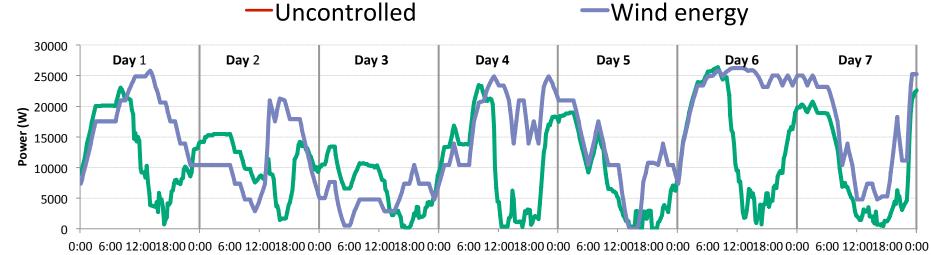
$$\min \quad \sum_{t=1}^{S} \left(w_t - \sum_{k=1}^{K} x_t^k \right)^2$$





Results: Uncontrolled BAU vs. Distributed





Distributed

—Wind energy

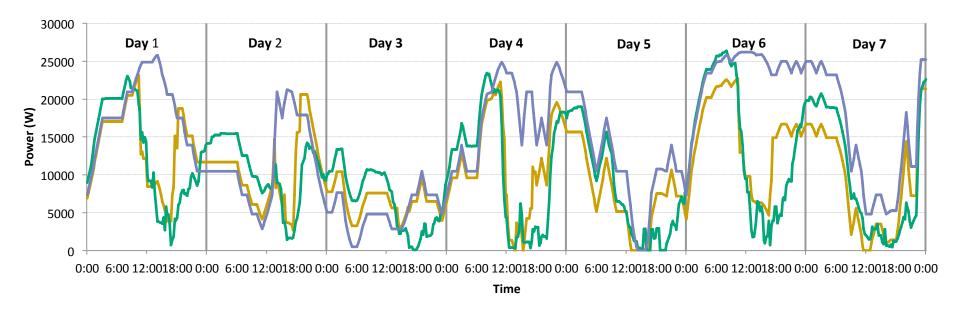




Time

Results: Distributed vs. Benchmark

Benchmark

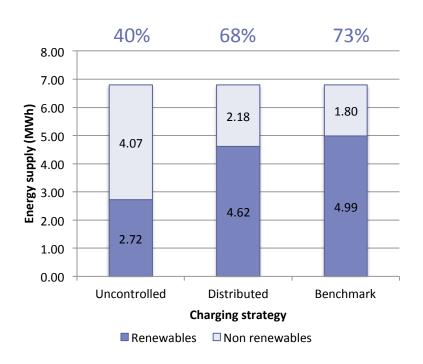


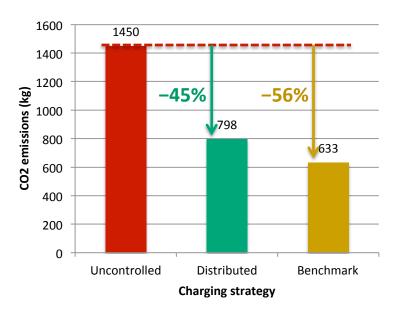




—Distributed —Wind energy

Results: Energy Mix





Renewables: 7.4 CO2 g/kWh Non Renewables: 351.0 CO2 g/kWh

- Total energy consumption ≈ 6.8 MWh
- Substantial increase in the use of renewable energy
- Reduced CO₂ emissions





Conclusions

- Objective: balance wind energy supply with electric vehicle charging demand
- Method: Distributed coordination algorithm in which participants exchange virtual prices and energy schedules
- Performance: Distributed coordination significantly better than BAU, close to "ideal world" benchmark
 - Increased usage of renewable energy sources
 - Reduction of CO₂ emissions





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K. De Craemer, **"Ch.3: Algorithms for demand response of electric vehicles"**, in: "Event-Driven Demand Response for Electric Vehicles in Multi-aggregator Distribution Grid Settings", Ph.D. Thesis, KU Leuven, Jul. 2014

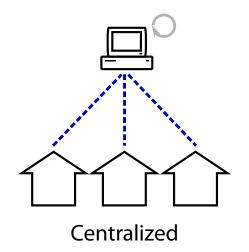


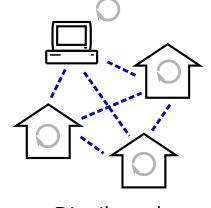


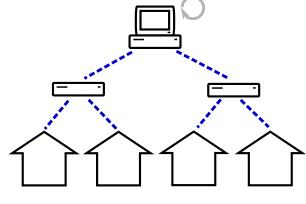


Strategies for DR

Optimization







Distributed

Game theory

Distributed

optimization

Aggregate & dispatch

- Approximate DP
- Stochastic programming
- Iterative local search to solve (M)ILP

...

- Ranking, MPC

 - State-bin modeling
 - Market-based

•

E.g., dual decomposition

...

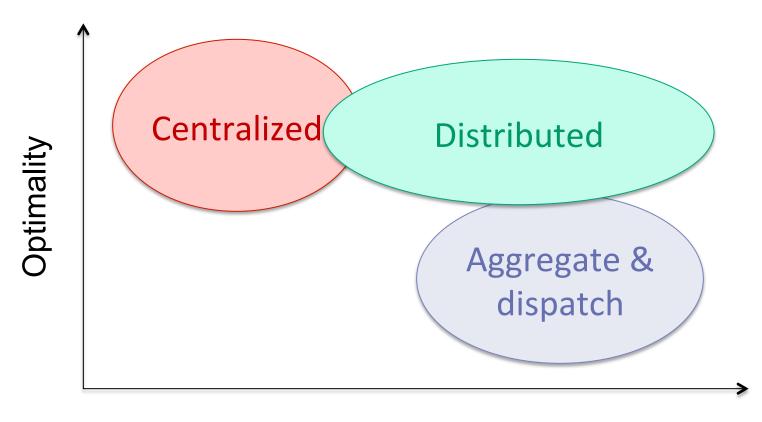
Figure adapted from [K. De Craemer, 2014]







Strategies for DR: Scalability vs optimality



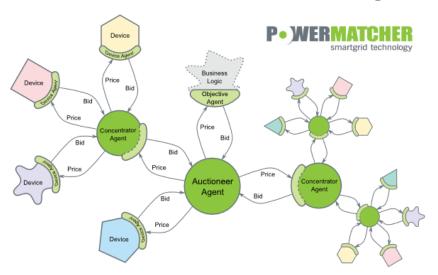
Cluster size





Multi-agent systems (MAS)

- Agent
 - = takes independent action to achieve its design objectives
 - ≠ told explicitly what to do exactly
- Multi-agent
 - = interacting agents, via message exchange
 - = cooperating/coordinating/negotiating agents
- Example: PowerMatcher = market-based agent system







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K. Mets, T. Verschueren, C. Develder, T. Vandoorn and L. Vandevelde, "Integrated Simulation of Power and Communication Networks for Smart Grid Applications", in Proc. 16th IEEE Int. Workshop Computer Aided Modeling, Analysis and Design of Commun. Links and Netw. (CAMAD 2011), Kyoto, Japan, 10-11 Jun. 2011, pp. 61-65. doi:10.1109/CAMAD.2011.5941119

K. Mets, J. Aparicio and C. Develder, "Combining power and communication network simulation for cost-effective smart grid analysis", IEEE Commun. Surveys Tutorials, Vol. PP, 2014, pp. 1-26.. doi:10.1109/SURV.2014.021414.00116





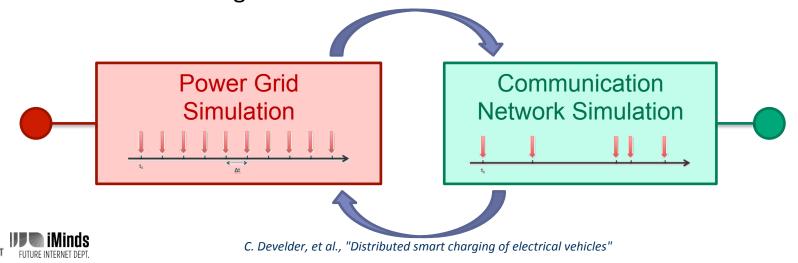


Problem Statement

- Simulators are already used in the two domains:
 - Communication network engineering
 - Power engineering



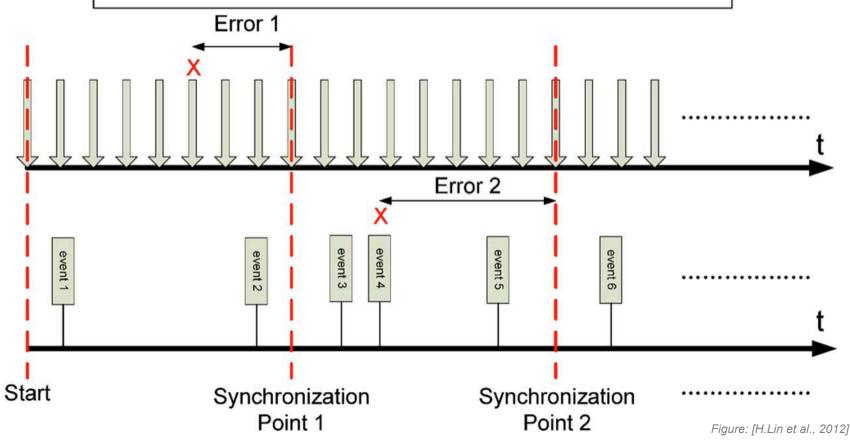
- In a co-simulation approach, power & communication are loosely coupled
 - Requires careful synchronisation
 - Drawback: no integration of tools





Challenge for co-simulation: Synchronisation

- Stands for a round of power system dynamic simulation
- Stands for a communication network event



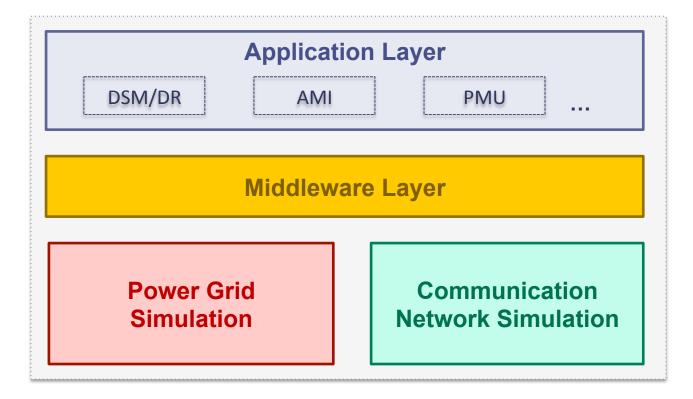




Our solution

Integrated (combined) power grid and communication network simulation

→ Large scale smart grid simulations







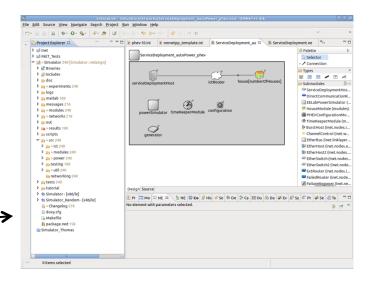
OMNeT++

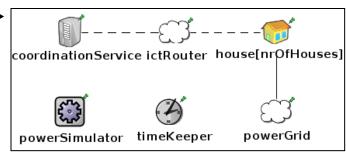
Discrete Event Simulator:

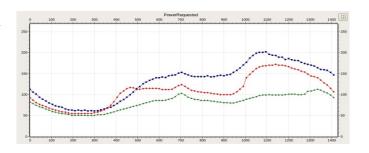
- Modular, Scalable, Cluster support
- Models for communication networks
- Integrated in Eclipse –
- Random Data Generation
- Graphical representations
- Data logging, presentation, processing, etc.
- Open source
- ...

• Custom Components:

- Electric components: loads, generators, etc.
- ICT components: smart devices, coordination services, ...











Power Flow Simulator

- Support for radial distribution grid topologies.
- Model based on Fast Harmonic Simulation Method [1].
- Model implemented in MATLAB and integrated in simulator.
- Uses an Iterative forward/backward sweep method:

INPUT:

- Power demand (Watt) at each node at time t.
- Phase to which each node is connected

LOOP:

Backward sweep

- Determines currents in every node, based on known voltages in each node.
- Currents in all network branches are determined.

Forward sweep

Determines voltage at every node

<u>Compare voltages</u> with the voltages in the previous iteration.

If difference below a certain threshold:

Stop iterations.

Else

Continue iterations.

[1] L. Degroote, L. Vandevelde, and B. Renders, "Fast harmonic simulation model for the analysis of network losses with converter-connecter distributed generation", *Electric Power System Research*, vol. 80, pp. 1332-1340, 2010.







Communication Network Models

INET Framework:

Open source communication network simulation package for OMNeT++

Layer	Protocol
Transport	TCP
	UDP
Network	IPv4
	IPv6
Link	Ethernet
	802.11 (WiFi)
	PPP



Implemented as OMNeT++ modules.

→ Simulate different technologies

... or basic OMNet++ message framework:

No specific protocol or physical layers are simulated

→ Reduced overhead compared to INET





Simulator Configuration

Nodes

Implementation of ICT and/or power components

The nodes and interconnections between nodes form the power grid and communication network that form the simulated smart grid.

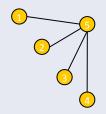
Network Description File

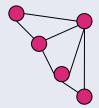
List of used modules

- Electrical network model
- Communication network
- Smart devices

Topology

- Electrical
- Communication





INI File

Node parameters

- Node type
- Generator capacity
- Battery charge rate
- ...

Simulation parameters

- # houses, # devices, ...
- Communication technologies
- Device properties
- Control algorithms

Defining input files

• E.g., load profiles







Outline

- 1. Introduction
- 2. Example 1: Peak shaving
- 3. Example 2: Wind balancing
- 4. DR algorithms for EV charging
- 5. Tools to study EV charging
- 6. Wrap-up





Outline

- 1. Introduction
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Take-away points

- Challenges for the grid:
 - Increased renewable energy sources + distributed into LV/MV network
 - Increase of load (e.g., EVs)
- Penetration of EVs:
 - HEV vs BEV
 - Need for peak shaving and/or balancing
 - Benefit of V2G?
- DR algorithms
 - Distributed vs centralized
 - Agent-based systems





Future & ongoing work

- Communication network architecture
 - C-DAX concept: generic smart grid middleware
 - Hierarchical architecture, e.g., using data aggregators to reduce communication overhead
 - Communication network requirements, impact of communication problems, etc.
- Algorithm development
 - Stochastic behavior
 - Multiple balancing zones
 - Vehicle-to-grid support





Thank you ... any questions?







Thank you ... any questions?

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